Accepting Hybrid Networks of Evolutionary Processors with Special Topologies and Small Communication

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Starting from the fact that complete Accepting Hybrid Networks of Evolutionary Processors allow much communication between the nodes and are far from network structures used in practice, we propose in this paper three network topologies that restrict the communication: star networks, ring networks, and grid networks. We show that ring-AHNEPs can simulate 2-tag systems, thus we deduce the existence of a universal ring-AHNEP. For star networks or grid networks, we show a more general result; that is, each recursively enumerable language can be accepted efficiently by a star- or grid-AHNEP. We also present bounds for the size of these star and grid networks. As a consequence we get that each recursively enumerable can be accepted by networks with at most 13 communication channels and by networks where each node communicates with at most three other nodes.

1 Introduction

Motivated by parallel architectures, networks of language processors were introduced by E. Csuhaj-Varjú and A. Salomaa in [3]. They consist of processors which process a word according to some rewriting rules and communicate by sending the words to other nodes via filters. Starting from finite sets of word in the nodes a language is generated consisting of all nodes occurring in some distinguished node. In [2] networks with evolutionary processors were introduced where the rewriting rules are restricted to substitutions, deletions and insertions of letters which describe the local mutations known from biology. It was shown by them that networks with evolutionary networks can generate all recursively enumerable languages.

In [9], an accepting counterpart for networks with evolutionary processors was introduced. A distinguished node contains a word which is processed by the network and accepted if a word occurs in another distinguished node. It was shown that any recursively enumerable language can be accepted by accepting networks with evolutionary processors (abbreviated by AHNEP). Moreover, AHENPs with seven nodes are sufficient to accept any recursively enumerable language (see [1]), while AHNEPs with ten nodes are sufficient to simulate efficiently a nondeterministic Turing machine (see [6]).

In most of the papers in the area, AHNEPs with an underlying complete graph were considered. As far as universality and computational and descriptional complexity results are concerned, it seems a natural decision to keep the shape of the network the same, no matter what language the network accepts; otherwise the different characterizations that one may obtain can be considered as not being uniform. However, complete graphs have two disadvantages.

 Complete graphs are very different from network structures which occur in practice where very often stars, grids, rings, backbones etc. are used.

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A complete graph has many edges which correspond to communication channels. If one is interested
in a small amount of communication, one has to consider other structures for the underlying graphs.

In this paper we consider AHNEPs where the underlying graph is a star or a grid or a ring. Regarding these structures one can expect to get also networks with a small number of communication channels, since the number of edges in a star is its number of nodes minus 1 and the maximal number of edges connected with a node is restricted by 4 in a grid.

We show that each recursively enumerable language can be accepted efficiently by a star or grid-AHNEP. Furthermore, we prove that the number of processors used in these star networks is bounded by 13 and that of grid networks by 52. As a consequence we get that each recursively enumerable language can be accepted by networks with at most 12 edges and by networks where each node communicates with at most three other nodes. In the case of ring networks, we only have a weaker result. Ring-AHNEPs can simulate 2-tag systems, thus we deduce the existence of a universal ring-AHNEP. Moreover, ring-AHNEPs are able to solve NP-complete problems in polynomial time.

2 Basic definitions

We start by summarizing the notions used throughout the paper; for all unexplained notions the reader is referred to [14, 5, 11, 13].

An alphabet is a finite and nonempty set of symbols. The set of all words over V is denoted by V^* and the empty word is denoted by ε . The length of a word x is denoted by |x| while alph(x) denotes the minimal alphabet W such that $x \in W^*$. For a word $x \in W^*$, x^r denotes the reversal of the word.

We say that a rule $a \to b$, with $a, b \in V \cup \{\varepsilon\}$ is a *substitution rule* if both a and b are not ε ; it is a *deletion rule* if $a \neq \varepsilon$ and $b = \varepsilon$; it is an *insertion rule* if $a = \varepsilon$ and $b \neq \varepsilon$. The set of all substitution, deletion, and insertion rules over an alphabet V are denoted by Sub_V , Del_V , and Ins_V , respectively.

Given a rule σ as above and a word $w \in V^*$, we define the following *actions* of σ on w:

• If
$$\sigma \equiv a \rightarrow b \in Sub_V$$
, then $\sigma^*(w) = \begin{cases} \{ubv : \exists u, v \in V^* \ (w = uav)\}, \\ \{w \mid w \text{ does not contain } a\}, \end{cases}$

• If
$$\sigma \equiv a \rightarrow \varepsilon \in Del_V$$
, then $\sigma^*(w) = \begin{cases} \{uv : \exists u, v \in V^* \ (w = uav)\}, \\ \{w \mid w \text{ does not contain } a\}, \end{cases}$

$$\sigma^r(w) = \begin{cases} \{u : w = ua\}, \\ \{w \mid w \text{ contains no } a\}, \end{cases}$$

$$\sigma^l(w) = \begin{cases} \{v : w = av\}, \\ \{w \mid w \text{ contains no } a\}, \end{cases}$$

• If
$$\sigma \equiv \varepsilon \rightarrow a \in Ins_V$$
, then $\sigma^*(w) = \{uav : \exists u, v \in V^* (w = uv)\}, \ \sigma^r(w) = \{wa\}, \ \sigma^l(w) = \{aw\}.$

We say that $\alpha \in \{*, l, r\}$ expresses the way of applying a deletion or insertion rule to a word, namely at any position $(\alpha = *)$, in the left $(\alpha = l)$, or in the right $(\alpha = r)$ end of the word, respectively. For every rule σ , action $\alpha \in \{*, l, r\}$, and $L \subseteq V^*$, we define the α -action of σ on L by

$$\sigma^{\alpha}(L) = \bigcup_{w \in L} \sigma^{\alpha}(w).$$

Given a finite set of rules M, we define the α -action of M on a word w and a language L by

$$M^{\alpha}(w) = \bigcup_{\sigma \in M} \sigma^{\alpha}(w)$$
 and $M^{\alpha}(L) = \bigcup_{w \in L} M^{\alpha}(w)$,

respectively. In what follows, we shall refer to the rewriting operations defined above as *evolutionary* operations since they may be viewed as linguistic formulations of local DNA mutations.

For two disjoint subsets P and F of an alphabet V and a word w over V, we define the predicates

$$\begin{array}{lcl} \varphi^{(s)}(w;P,F) & \equiv & P \subseteq alph(w) \land F \cap alph(w) = \emptyset, \\ \varphi^{(w)}(w;P,F) & \equiv & (P = \emptyset \lor alph(w) \cap P \neq \emptyset) \land F \cap alph(w) = \emptyset. \end{array}$$

The construction of these predicates is based on random-context conditions defined by the two sets P permitting contexts/symbols and F of forbidding contexts/symbols. Informally, the first condition requires that all permitting symbols are present in w and no forbidding symbol is present in w, while the second one is a weaker variant of the first, requiring that – for an non-empty set P – at least one permitting symbol appears in w and no forbidding symbol is present in w.

For every language $L \subseteq V^*$ and $\beta \in \{(s), (w)\}$, we define:

$$\varphi^{\beta}(L, P, F) = \{ w \in L \mid \varphi^{\beta}(w; P, F) \}.$$

An evolutionary processor over V is a tuple (M, PI, FI, PO, FO), where:

- $-(M \subseteq Sub_V)$ or $(M \subseteq Del_V)$ or $(M \subseteq Ins_V)$ and
- -PI,FI,PO,FO ⊂ V.

M is a set of substitution, deletion or insertion rules over the alphabet *V*. The set *M* represents the set of evolutionary rules of the processor. As one can see, a processor is "specialized" in only one evolutionary operation. *PI* and *FI* are the *input* permitting/forbidding contexts of the processor, while *PO* and *FO* are the *output* permitting/forbidding contexts of the processor. Informally, the permitting input/output contexts are the set of symbols that should be present in a word, when it enters/leaves the processor, while the forbidding contexts are the set of symbols that should not be present in a word in order to enter/leave the processor.

We denote the set of evolutionary processors over V by EP_V . Note that the evolutionary processor described here is a mathematical concept similar to that of an evolutionary algorithm, both being inspired from Darwinian evolution. The rewriting operations we have considered might be interpreted as mutations and the filtering process described above might be viewed as a selection process. Recombination is missing but it was asserted that evolutionary and functional relationships between genes can be captured by taking only local mutations into consideration [15]. Furthermore, we are not concerned here with a possible biological implementation of these processors, though this is a matter of great importance.

An accepting hybrid network of evolutionary processors (AHNEP for short) is a 7-tuple $\Gamma = (V, U, G, \mathcal{N}, \alpha, \beta, x_I, x_O)$, where

- V and U are the input and network alphabets, respectively, with $V \subseteq U$.
- $-G = (X_G, E_G)$ is an undirected graph, with the set of nodes X_G and the set of edges E_G .
- $-\mathscr{N}: X_G \longrightarrow EP_U$ is a mapping which associates with each node $x \in X_G$ an evolutionary processor $\mathscr{N}(x) = (M_x, PI_x, FI_x, PO_x, FO_x)$.
- $-\alpha: X_G \longrightarrow \{*, l, r\}; \alpha(x)$ gives the action mode of the rules of node x on the words existing in that node.
- $-\beta: X_G \longrightarrow \{(s), (w)\}$ defines the type of the *input/output filters* of a node.
- $-x_I$ and $x_O \in X_G$ is the *input node*, and the *output node*, respectively, of the AHNEP.

G is called the *underlying graph* of the network. We denote an undirected edge between nodes x and y by (x,y).

For every node $x \in X_G$, the following filters are defined:

input filter:
$$\rho_x(\cdot) = \varphi^{\beta(x)}(\cdot; PI_x, FI_x),$$
 output filter: $\tau_x(\cdot) = \varphi^{\beta(x)}(\cdot; PO_x, FO_x).$

That is, $\rho_x(w)$ (resp. τ_x) indicates whether or not the word w can pass the input (resp. output) filter of x. More generally, $\rho_x(L)$ (resp. $\tau_x(L)$) is the set of words of L that can pass the input (resp. output) filter of x.

We say that the number of nodes present in the AHNEP Γ is the size of Γ .

A configuration of an AHNEP Γ as above is a mapping $C: X_G \longrightarrow 2^{V^*}$ which associates a set of words with every node of the graph. A configuration may be understood as the sets of words which are present in any node at a given moment. A configuration can change either by an evolutionary step or by a communication step.

When changing by an evolutionary step each component C(x) of the configuration C is changed in accordance with the set of evolutionary rules M_x associated with the node x and the way $\alpha(x)$ of applying these rules. Formally, we say that the configuration C' is obtained in an *evolutionary step* from the configuration C, written as $C \Longrightarrow C'$, if and only if $C'(x) = M_x^{\alpha(x)}(C(x))$ for all $x \in X_G$.

When changing by a communication step, each node processor $x \in X_G$ sends one copy of each word it has, which is able to pass the output filter of x, to all the node processors connected to x and receives all the words sent by any node processor connected with x providing that they can pass its input filter. Formally, we say that the configuration C' is obtained in a *communication step* from configuration C, written as $C \vdash C'$, if and only if

$$C'(x) = (C(x) - \tau_x(C(x))) \cup \bigcup_{(x,y) \in E_G} (\tau_y(C(y)) \cap \rho_x(C(y)))$$

for all $x \in X_G$. Note that words which leave a node x are eliminated from that node; if they cannot pass the input filter of any node connected with x, they are lost.

Let Γ be an AHNEP, the computation of Γ on the input word $w \in V^*$ is a sequence of configurations $C_0^{(w)}, C_1^{(w)}, C_2^{(w)}, \ldots$, where the initial configuration $C_0^{(w)}$ of Γ is defined by $C_0^{(w)}(x_I) = \{w\}$ and $C_0^{(w)}(x) = \emptyset$ for all $x \in X_G$, $x \neq x_I$, and $C_{2i}^{(w)} \Longrightarrow C_{2i+1}^{(w)}$ and $C_{2i+1}^{(w)} \vdash C_{2i+2}^{(w)}$ holds for all $i \geq 0$. By the previous definitions, each configuration $C_i^{(w)}$ is uniquely determined by the configuration $C_{i-1}^{(w)}$. Thus each computation in an AHNEP is deterministic. A computation as above immediately halts if one of the following two conditions holds:

- (i) There exists a configuration in which the set of words existing in the output node x_O is non-empty. In this case, the computation is said to be an *accepting computation*.
- (ii) There exist two identical configurations obtained either in consecutive evolutionary steps or in consecutive communication steps. If this condition is fulfilled, but condition (i) is not fulfilled, then we say that the computation is a *rejecting computation*.

In the aforementioned cases the computation is said to be finite.

The *language accepted* by Γ is

$$L_a(\Gamma) = \{ w \in V^* \mid \text{ the computation of } \Gamma \text{ on } w \text{ is an accepting one} \}.$$

We say that an AHNEP Γ decides the language $L \subseteq V^*$, and write $L(\Gamma) = L$ if and only if $L_a(\Gamma) = L$ and the computation of Γ halts on every $x \in V^*$.

Let Γ be an AHNEP deciding the language L. The *time complexity* of the finite computation $C_0^{(x)}$, $C_1^{(x)}$, $C_2^{(x)}$, ... $C_m^{(x)}$ of Γ on $x \in L$ is denoted by $Time_{\Gamma}(x)$ and equals m. The time complexity of Γ is the function from $\mathbf N$ to $\mathbf N$,

$$Time_{\Gamma}(n) = \sup\{Time_{\Gamma}(x) \mid |x| = n\}.$$

In this paper we consider networks of evolutionary processors with special underlying graphs.

The first network topology we consider will be called *star*. An *n*-star, with $n \ge 1$, is a graph having n+1 vertices, say x_1, x_2, \ldots, x_n and C, and the edges (x_i, C) for $1 \le i \le n$. One can easily imagine such a network as a center node C connected to n other nodes that are placed around it. An AHNEP having as underlying graph a n-star for some $n \ge 1$, will be called \star AHNEP (or star-AHNEP).

The second network topology we consider will be called *ring*. An *n*-ring, with $n \ge 2$, is a graph having n + 1 nodes, say x_1, x_2, \ldots, x_n (called the ring-nodes) and x_0 (called the center-node), and the edges $(x_i, x_{(i \mod n)+1})$ and (x_i, x_0) for $1 \le i \le n$. One can easily imagine such a network as n nodes disposed on a ring, each two consecutive nodes being connected, with another node right in the middle of the ring, connected to all the others (however, this node will have a different use than the central node in \star AHNEPs). An AHNEP having as underlying graph such a ring and verifying the property that the central node x_0 of the graph is always the output node of the network will be called a \circ AHNEP (or ring-AHNEP). In this way, the computation is done by the ring-nodes, and it stops only when a string enters the center-node (who does not have any other effect on the computation). It is worth to remark that the topology we use in this case is called ring because the nodes that perform the computation form a ring; putting the output node on the same ring would lead to the disconnection of this ring and its transformation into a chain (because the output node cannot be of any use in the computation, and, moreover, once a string enters in this node, the whole process stops).

The third network topology we consider will be called a grid. An (m,n)-grid is a graph consisting of $m \cdot n$ nodes (i,j), $1 \le i \le m$ and $1 \le j \le n$, and having the edges ((i,j),(i+1,j)) and ((i,j),(i,j+1)) for $1 \le i \le m-1$ and $1 \le j \le n-1$, ((m,j),(m,j+1)) for $1 \le j \le n-1$ and ((i,n),(i+1,n)) for $1 \le i \le m-1$. Obviously, an (m,n)-grid is isomorphic to the finite subgrid of the grid $\mathbb{Z} \times \mathbb{Z}$ of height n-1 and width m-1. An AHNEP having as an underlying graph an (m,n)-grid for some $n \ge 1$ and $m \ge 1$ will be called a grid-AHNEP.

Ring and star networks were also approached in [2] but in the context of generating NEPs with regular filters. However, the results and proofs presented in that paper cannot be applied to get similar results for AHNEPs with random context filters.

3 Star-AHNEPs

We start with a further characterization of the class of recursively enumerable (recursive) languages by networks where the underlying graphs are complete graphs with loops.

Theorem 1. For any recursively enumerable (recursive) language L accepted (decided) by a Turing machine there exists a AHNEP Γ of size 12 such that the underlying graph is a complete graph which additionally has a loop for each node, and it accepts (decides) L.

Moreover, if $L \in NTIME(f(n))$, then $Time_{\Gamma}(n) \in \mathcal{O}(f(n))$ (where the constant hidden by the \mathcal{O} notation depends on the Turing machine accepting, respectively deciding, L).

By limitedness of space we omit the proof. Note, however, that it follows mainly the proof of the main result in [6].

The main result that we show in this section is the following theorem.

Theorem 2. For any recursively enumerable (recursive) language L accepted (decided) by a Turing machine there exists a \star AHNEP Γ of size 13 accepting (deciding) L. Moreover, if $L \in NTIME(f(n))$, then $Time_{\Gamma}(n) \in \mathcal{O}(f(n))$ (where the constant hidden by the \mathcal{O} notation depends on the Turing machine accepting, respectively deciding, L).

Proof. The proof is based on the following simple remark. Any AHNEP having as underlying graph a complete graph with loops (i.e., there is an edge between every two nodes, and an edge connecting each node to itself) can be canonically transformed into a \star AHNEP. Assume that the initial AHNEP has n nodes. In the new \star AHNEP we will have all the nodes of the initial network (with the same set of rules and filters), and a new one, called here C (with no rules and filters allowing any word to pass the filters). Every node of the former network is connected to C, and no other edges exist. It is not hard to see that the new network accepts (decides) exactly the same language that is accepted (decided) by the initial one. Indeed, the evolution steps are performed by the same nodes in both networks, but the communication that takes place between two nodes of the initial network is done here via the C node (that does not modify in any way the string). In conclusion, we have a way to transform AHENPs that have as underlying network a complete graph with loops into \star AHNEPs, by adding only one node. Moreover, the \star AHENP that we obtain simulates efficiently a computation of the initial AHENP: it makes at most two times the number of steps as in the initial computation of the complete AHNEP with loops.

Now the result follows immediately from Theorem 1. \Box

Using the same idea we can prove the following statement.

Theorem 3. For every 2-tag system $T = (V, \phi)$ there exists a $\star AHNEP \Gamma$ of size 7 such that $L(\Gamma) = \{w \mid T \text{ halts on } w\}$.

We stress out that this result does not improve, neither from the descriptive complexity point of view nor from the computational complexity point of view, the result in Theorem 2. Indeed, not every recursively enumerable language can be accepted by 2-tag systems, thus Theorem 3 does not imply that *AHNEPs with 7 nodes are complete; moreover, although 2-tag systems efficiently simulate deterministic Turing machines, via cyclic tag systems (see, e.g., [16]), the previous result does not allow us to infer a bound on the size of the networks accepting in a computationally efficient way all recursively enumerable languages.

However, from Theorem 3, and the fact that 2-tag systems are universal, we derive the following fact.

Corollary 1. There exists a universal ★AHNEP with 7 nodes.

In this respect we discuss a bit on the importance of the descriptive complexity bounds shown here: we know that *AHNEPs with 13 nodes accept efficiently recursive enumerable languages, and, as a simple consequence, we have that NP equals the class of languages accepted in polynomial time by *AHNEPs with 13 nodes. Why is such a result interesting, since, after all it is a rather clear fact that AHNEPs simulate the nondeterminism (of Turing machines, for instance) by working with strings with multiplicities? Our answer to this question would be that the result is interesting because it shows that a special class of AHNEPs (with a constant number of nodes, and a special topology) has these good computational properties; it is not clear if AHNEPs with less nodes have the same properties (we were not able neither to get a lower bound neither to show an equivalent of Theorem 2 for smaller AHNEPs).

If we are interested in a small number of communication channels in the net, then we have the following result which follows immediately from Theorem 2.

Corollary 2. For any recursively enumerable language L, there is an AHNEP Γ with at most 12 edges in the underlying graph such that $L = L_a(\Gamma)$.

4 Grid-AHNEPS

In this section we prove that grid-AHNEPS have the same power as *AHNEPs. More precisely, we show the following the following statements.

Theorem 4. Let $\Gamma = (V, U, G, \mathcal{N}, \alpha, \beta, x_i, x_o)$ be a network of size n where the underlying graph $G = (X_G, E_G)$ is a complete graph with loops. Then there is a (4, n+1)-grid-AHNEP Γ' such that $L_a(\Gamma') = L_a(\Gamma)$.

Proof. Let $\Gamma = (V, U, G, \mathcal{N}, \alpha, \beta, x_i, x_o)$ be a network where the underlying graph $G = (\{1, 2, ..., n\}, E_G)$ is a complete graph with loops (x, x) for any node $x \in X_G$. Then we construct the grid-AHNEP

$$\Gamma' = (V', U, G', \mathcal{N}', \alpha', \beta', (3, x_i + 1), (3, x_o + 1))$$

where G' is the grid consisting of the nodes (i, j) with $1 \le i \le n + 1$, $1 \le j \le 5$,

$$V' = V \cup \bigcup_{i=0}^{n} \{X_{i}, X'_{i}Y_{i}, Y'_{i}\},$$

$$\mathcal{N}'(1,1) = (\{X_{0} \to X_{1}\}, \{X_{0}\}, \emptyset, \emptyset, \emptyset), \ \alpha'(1,1) = *, \ \beta'(1,1) = (w),$$

$$\mathcal{N}'(1,i+1) = (\{X_{i} \to X_{i+1}, X_{i} \to X'_{i}\}, \{X_{i}\}, \emptyset, \emptyset, \emptyset), \ \alpha'(1,i+1) = *,$$

$$\beta'(1,i+1) = (w) \text{ for } 1 \leq i \leq n-1,$$

$$\mathcal{N}'(1,n+1) = (X_{n} \to X'_{n}\}, \{X_{n}\}, \emptyset, \emptyset, \emptyset), \ \alpha'(1,n+1) = *, \ \beta'(1,n+1) = (w),$$

$$\mathcal{N}'(2,1) = (\{Y \to \varepsilon\}, \{X_{0}\}, \emptyset, \emptyset, \{Y\}\}), \ \alpha'(2,1) = *, \beta'(2,1) = (w),$$

$$\mathcal{N}'(2,i+1) = (\{X'_{i} \to \varepsilon\}, \{X'_{i}\}, \emptyset, \emptyset, \emptyset), \ \alpha'(2,i+1) = *, \beta'(2,i+1) = (w)$$

$$\text{for } 1 \leq i \leq n-1,$$

$$\mathcal{N}'(3,1) = (\{Y' \to X_{0}\}, \{Y'\}, \emptyset, \emptyset, \emptyset), \ \alpha'(3,1) = *, \ \beta'(3,1) = (w),$$

$$\mathcal{N}'(3,i+1) = (M_{i}, PI_{i}, FI_{i} \cup \{Y, X_{0}\}, PO_{i}, FO_{i}), \ \alpha'(3,i+1) = \alpha(i),$$

$$\beta'(3,i+1) = \beta(i) \text{ for } 1 \leq i \leq n,$$

$$\mathcal{N}'(4,1) = (\{Y \to Y'\}, \{Y\}, \emptyset, \emptyset, \emptyset), \ \alpha'(4,1) = *, \ \beta'(4,1) = (w)$$

$$\mathcal{N}'(4,i+1) = (\{\varepsilon \to Y\}, \emptyset, \{Y'\}, \emptyset, \emptyset), \ \alpha'(4,n+1) = *, \ \beta'(4,n+1) = (w)$$

We now consider the case that a word $z \in V^*$ is contained in the node (3, i+1) for some $i, 1 \le i \le n$. Note that in the initial configuration the word w is in $(3, x_i + 1)$. Then we apply an evolutionary rule of M_i and obtain a word $z' \in V^*$. This can be communicated to node (3, i) and (3, i+2) (if $i \ge 2$ and $i \le n-1$, respectively), if the corresponding filters allow the communication. Since G is a complete graph, these communications can be done in Γ , too. Furthermore, z' can be communicated to (4, i+1). The application of $\varepsilon \to Y$ gives a word $z_1'Yz_2'$ with $z' = z_1'z_2'$. Now the word can only move in the fourth row of nodes, i. e., only to nodes (4, j) with $1 \le j \le n+1$. As long as nodes (4, j) with $1 \le j \le n+1$ are entered, a further Y is inserted somewhere in the word. If (4, 1) is entered, then one Y is changed to Y' and the word has to move to node (3, 1) where Y' is replaced by Y_0 . Now the word has to move to node (2, 1) where the remaining Ys in the word are deleted. The obtained word $y'_1 = y'_2 = y'_3 = y'_4 = y'_4$

whereas the application of $X_i \to X_i'$ results in a vertical move to (2,k+1). One of these possibilities have to be chosen in each node (1,k+1). Let us assume that we choose the vertical direction for some j. Then we obtain $z_1''X_j'z_2''$ in node (2,j+1). The application of $X_j' \to \varepsilon$ leads to $z_1''z_2'' = z'$ in node (3,j+1), if the filters of (3,j+1) allow the passing. Altogether we have performed a move from (i,3) to (j,3) and taking into consideration the filters the move from i to j can be done in Γ , too. Therefore we have $L_a(\Gamma') \subseteq L_a(\Gamma)$.

Let us denote by $s_{i,j}$ the sequence of nodes

$$(4, i+1), (4, i), (4, i-1), \dots, (4, 1), (3, 1), (2, 1), (1, 1), (1, 2), \dots (1, j+1), (2, j+1).$$

Let $w \in L_a(\Gamma)$. Then there is a sequence $x_i, t_1, t_2, \dots, t_n, x_o$ of nodes which are entered in this succession in the computation on w starting in x_i . Then we have in Γ' a computation on w which follows the sequence

$$(3, x_i + 1), s_{x_i,t_1}, (3, t_1 + 1), s_{t_1,t_2}, (3, t_2 + 1), s_{t_2,t_3} \dots (3, t_n + 1)s_{t_n,x_o}(3, x_o + 1).$$

Thus $w \in L_a(\Gamma')$, too. Therefore we also have $L_a(\Gamma) \subseteq L_a(\Gamma')$.

The following corollary follows immediately from Theorem 1.

Corollary 3. For any recursively enumerable language L, there is a grid-AHNEP Γ' of size at most 52 such that $L = L_a(\Gamma')$.

Obviously, any node of a grid-AHNEP is connected with at most four other nodes. Thus, in order to accept recursively enumerable languages it is sufficient to consider graphs where each node has a degree ≤ 4 . This result is slightly improved by the following statement.

Theorem 5. For any recursively enumerable language L, there is a AHNEP Γ such that any node of the underlying graph has a degree ≤ 3 and $L_a(\Gamma) = L$.

Proof. We consider the network constructed in the proof of Theorem 4. It is obvious, that the edges ((i,j),(i+1,j)) for $1 \le j \le n$ and $2 \le j \le 3$ can be omitted without changing the accepted language. Clearly, after the cancellation the maximal degree of the nodes is three.

We do not know whether or not this result is optimal. If the degree is bounded by 2, one gets chains of nodes (see the remark following the definition of ring-AHNEPs). We leave as an open question whether AHNEPs with chains as underlying graphs are able to accept all recursively enumerable languages.

5 Ring-AHNEPs

With respect to ring-AHNEPs we were not able to prove that all recursively enumerable languages can be accepted. However, ring-AHNEPs also have a large power since they are able to simulate the universal 2-tag systems and to solve NP-complete problems in polynomial time.

Theorem 6. For every 2-tag system $T = (V, \phi)$ there exists a $\circ AHNEP \Gamma$ of size 9 such that $L(\Gamma) = \{w \mid T \text{ halts on } w\}$.

We omit the proof by limitedness of space.

Once again, we remark that this result does not give a characterization of what class of languages oAHNEPs accept. We only proved that for any recursively enumerable language one can construct a oAHNEP that accepts a special encoding of that language. Since 2-tag systems efficiently simulate deterministic Turing machines ([16]), the previous result also shows that oAHNEPs with 8 nodes also accept efficiently encodings of the languages accepted efficiently by deterministic Turing machines.

A direct consequence of Theorem 6 is the following assertion.

Corollary 4. There exists a universal oAHNEP with 9 nodes.

Finally, in this section we present an application of \circ AHNEPs to efficient problem solving. First, let us recall what we mean by solving a problem with AHNEPs (see [8]). A correspondence between decision problems and languages can be done via an encoding function which transforms an instance of a given decision problem into a word, see, e.g., [4]. We say that a decision problem P is solved in time O(f(n)) by AHNEPs if there exists a family $\mathscr G$ of AHNEPs such that the following conditions are satisfied:

- The encoding function of any instance p of P having size n can be computed by a deterministic Turing machine in time O(f(n)).
- For each instance p of size n of the problem one can effectively construct, in deterministic time O(f(n)), an AHNEP $\Gamma(p) \in \mathscr{G}$ which decides, again in time O(f(n)), the word encoding the given instance. This means that the word is decided if and only if the solution to the given instance of the problem is "YES". This effective construction is called an O(f(n)) time solution to the considered problem.

Satisfiability is perhaps the best studied NP-complete problem because one arrives at it from a large number of practical problems. It has direct applications in mathematical logic, artificial intelligence, VLSI engineering, computing theory, etc. It can also be met indirectly in the area of constraint satisfaction problems. We now mention a result on getting all solutions to an instance of 3CNF-SAT (see [4]) using \circ AHNEPs.

Theorem 7. The satisfiability of formulas given in 3-Conjunctive Normal Form can be solved in polynomial time by $\circ AHNEPs$.

The proof of this theorem is based on the following idea: for a formula over n variables, we generate in $\mathcal{O}(n^2)$ AHNEP-time all the possible assignments for these variables, in a massively parallel manner; then, if the formula is the conjunction of m clauses, we evaluate it, for all the possible assignments, in parallel, in $\mathcal{O}(m)$ time. By limitedness of space we omit the detailed proof. Also, the result of the previous theorem can be extended, using a very similar proof, for a class of AHNEPs with simpler underlying graph, namely AHNEPs in which every node has degree 2 (or, simpler, AHNEPs having as underlying graph a chain of nodes): the satisfiability of formulas given in 3-Conjunctive Normal Form can be decided in polynomial time by such AHNEPs.

Note that this result does not say much about the computational power of AHNEPs having the underlying graph a chain; we were not able to show that they can be used to accept arbitrary recursively enumerable languages (as we did for star-AHNEPs and grid-AHNEPs, in Theorem 2 and, respectively, Corollary 3), or even encodings of such languages (as we did for ring-AHNEPs, in Theorem 6). In fact, we only showed that for all the instances of the Satisfiability problem, having the same number of variables, one can construct an AHNEP that decides for which formulas there exists an assignment of the variables that make them true, and for which there is no such assignment; it is important to note that a new network should be constructed every time we have to check a formula with a different number of variables.

Besides showing the completeness of star-AHNEPs and grid-AHNEPs, Theorem 2 and Corollary 3 show that networks with star or grid topology can be used to efficiently solve NP-complete problems. However, as far as problem solving is concerned, we can easily adapt the constructions presented in [7, 8] to see how NP-complete problems can be solved efficiently by complete AHNEPs with loops; thus, we can easily obtain efficient solutions implemented on star-AHNEPs, or grid-AHNEPs, for the same problems.

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